

## Storage Management and Data Mining Problems in High Energy Physics Applications

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(<http://www.lbl.gov/DM.html>)

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## Data Organization and Indexing of Large High Energy Physics Data

Collaboration	# members /institutions	Date of first data	# events/year	total data volume/year-TB
STAR	350/35	1999	$10^7$ - $10^8$	300
PHENIX	350/35	1999	$10^9$	600
BABAR	300/30	1999	$10^9$	80
CLAS	200/40	1997	$10^{10}$	300
ATLAS	1200/140	2004	$10^9$	2000

STAR: Solenoidal Tracker At RHIC  
RHIC: Relativistic Heavy Ion Collider

## Data Organization for Efficient Retrieval of Very Large Datasets

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- **General problem area**
  - how to cluster data in physical storage according to expected access patterns
- **Observation**
  - on parallel disks: distribute clusters to maximize parallel reads
  - on tape storage: keep cluster together to minimize tape mounts

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## HENP Mass Storage Access

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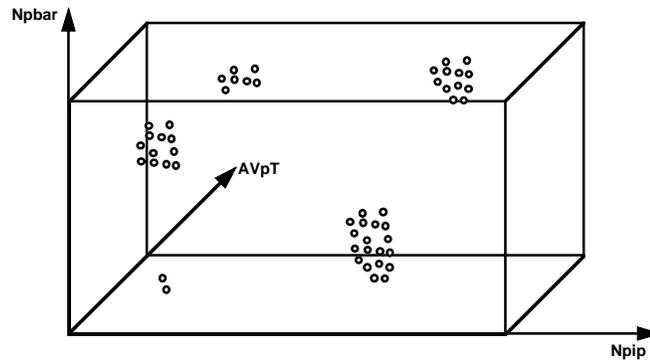
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- **After event reconstruction, event properties (features) are extracted: called “level 1 n-tuples”**
- **Number of properties is large (50-100)**
- **e.g. momentum, no. of pions, transverse energy**
- **Multidimensional space is highly skewed and sparse**
- **Need to access events based on partial properties specification (usually ranges)**
- **Problem: re-organize event clusters on mass storage according to the property space**

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# Clusters in the M-Dim property space

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## EXAMPLE OF EVENT PROPERTY VALUES

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I event 1	I Np(3) 24	R AVpT(1) 0.325951
I N(1) 9965	I Npbar(1) 94	R AVpT(2) 0.402098
I N(2) 1192	I Npbar(2) 12	R AVpTpip(1) 0.300771
I N(3) 1704	I Npbar(3) 24	R AVpTpip(2) 0.379093
I Npip(1) 2443	I NSEC(1) 15607	R AVpTpip(1) 0.298997
I Npip(2) 551	I NSEC(2) 1342	R AVpTpip(2) 0.375859
I Npip(3) 426	I NSECpip(1) 638	R AVpTkp(1) 0.421875
I Npim(1) 2480	I NSECpip(2) 191	R AVpTkp(2) 0.564385
I Npim(2) 541	I NSECpim(1) 728	R AVpTkm(1) 0.435554
I Npim(3) 382	I NSECpim(2) 206	R AVpTkm(2) 0.663398
I Nkcp(1) 229	I NSECkp(1) 3	R AVpTp(1) 0.651253
I Nkcp(2) 30	I NSECkp(2) 0	R AVpTp(2) 0.777526
I Nkcp(3) 50	I NSECkm(1) 0	R AVpTpbar(1) 0.399824
I Nkm(1) 209	I NSECkm(2) 0	R AVpTpbar(2) 0.690237
I Nkm(2) 23	I NSECp(1) 524	I NHIGHpT(1) 205
I Nkm(3) 32	I NSECp(2) 244	I NHIGHpT(2) 7
I Np(1) 255	I NSECpbar(1) 41	I NHIGHpT(3) 1
I Np(2) 34	I NSECpbar(2) 8	I NHIGHpT(4) 0
		I NHIGHpT(5) 0

54 Properties, eventually  $10^8$  events

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## Size of indexes for STAR

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- **Index size**
  - property space:  $10^8$  events x 200 bytes = 20 GB
  - index space:  $10^5$  clusters x 100 bytes = 10 MB  
(assume 1000 events/cluster)
- **Problem**
  - how to organize property space index

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## Main Tasks

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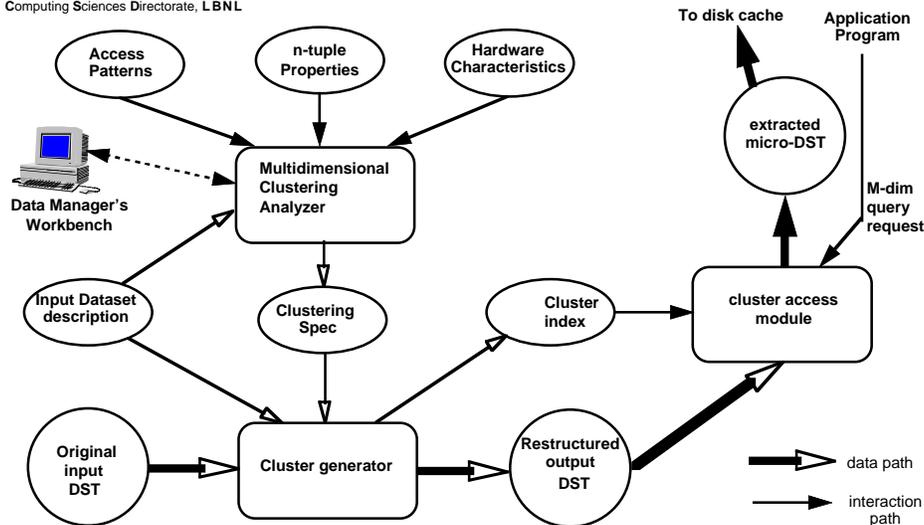
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- **Discover event clusters**
  - based on natural distribution - Data Mining
  - based on access patterns - consult physicists
  - simulate performance - data manager's workbench
- **Manage cluster access**
  - given a query, determine clusters to access,  
use multi-dimensional indexes to select events that qualify
- **Reorganize DST tapes according to clusters**
  - long process - done initially, then rarely
  - flow control - restart after interruption
- **Cache management**
  - determine if in cache, which incremental clusters to cache,  
which clusters to purge from cache

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## Events Clustering and Access: Main Components

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## Discover events clusters

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- **Top down approach**
  - partition each dimension into “bins” (e.g. 1-2 GEV, ..., 1-3 pions, ...)
  - select subset of dimensions based on physicist’s experience
  - analyze which events fall into the same “cell” (i.e. m-dim rectangles formed by the bins)
  - eliminate empty cells
  - combine cells to form similar size clusters
- **Assumption**
  - most queries are “range queries”

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## Discover events clusters

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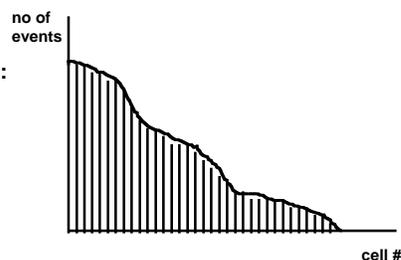
- **Bottom up approach**
  - select a subset of the dimensions (using physicist's hints initially)
  - partition space on each dimension successively
  - determine suitability of various indexing methods to high dimensionality and skewed distributions: K-D trees, Quad-trees, R-trees,...
  - Iterate for other dimension combinations
- **Consult physicists**
  - do cluster correlations matter?
  - get additional hints on preferred dimensions

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## Top Down Cell Management

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- **Assume: 7 dimensions, 10 bins each**
  - Number of cells:  $10^7$ , 4 byte counters
  - Number of bytes:  $4 \times 10^7$ , 40 MB
- **For e.g. small dataset 97% of cells are empty:**
  - store only populated cells
  - use hash tables to locate existing cells
  - use 2 bytes for bin\_id per property:  
ratio for p% full is:  $200 / (n+2)p$
  - No of bytes: for 7 dim, 3% => 5.4 MB
- **Sort cells by size (number of events)**

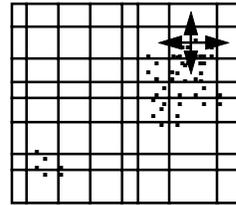


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## Cluster Identification

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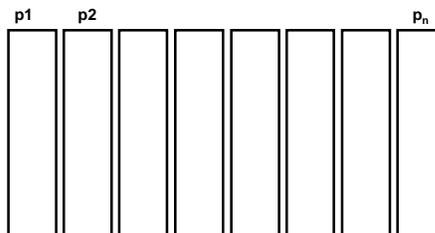
- **sort cells by size**
- **pick larger cell to start forming a cluster**
- **find all neighbors of “Manhattan distance” equal to 1**
- **include cells above a threshold**
- **iterate for all cells in cluster**
- **when no more cells above threshold, pick larger remaining cell and start forming a new cluster**
- **Display cluster distributions**



## indexing over all properties

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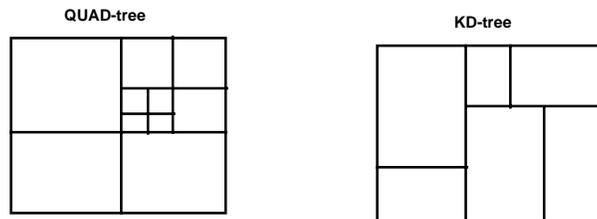
- **Assume 150 properties**
  - Any combination of range queries
  - want to compute number of events
- **possible solution: vertically partitioned file**
  - idea: touch only properties in queries
  - each partition  $10^8 \times 4$  bytes = 400 MB per partition
  - too expensive in space and time



# indexing over all properties

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- **other possible solutions**
  - partitioning MD space (KD-trees, n-QUAD-trees, ...)
  - for high dimensionality - either fanout or tree depth too large
  - e.g. symmetric n-QUAD-trees require  $2^{150}$  fanout
  - non-symmetric solutions are order dependent

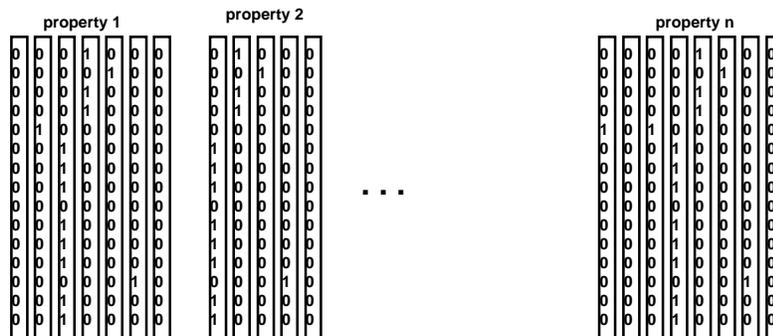


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# Bit-Sliced indexing

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- **partition each property into bins**
- **for each bin generate a bit vector**
- **compress each bit vector (run length encoding)**



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## Compression method

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**Uncompressed:**

0000000000001111000000000 .....000000100000000111111100000000 .... 000000

**Compressed:**

12, 16, 1016,1017,1025,2025

### Advantage:

Can perform: AND, OR, COUNT operations on compressed data

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## Bit-Sliced indexing

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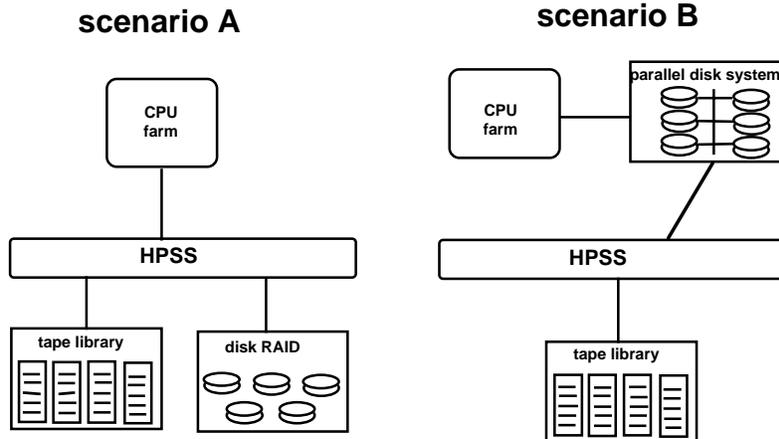
- **Estimated size**
  - 100 properties X 10 bins X  $10^8$  bits =  $10^{11}$  bits
  - compression factor (avg run length) = 1000
  - total size =  $10^8$  bits ~ 10 MB
- **Advantage**
  - only bit partitions need to be accessed (multiple bins per property are “or”ed, result for each property requested are “and”ed)
  - operations can be performed on compressed bit-slices
  - compressed bit-slices can be processed in parallel
- **Disadvantage**
  - results can be given on bin boundaries only

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# Cache Management

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## Hardware scenarios



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# Cache Management Issues

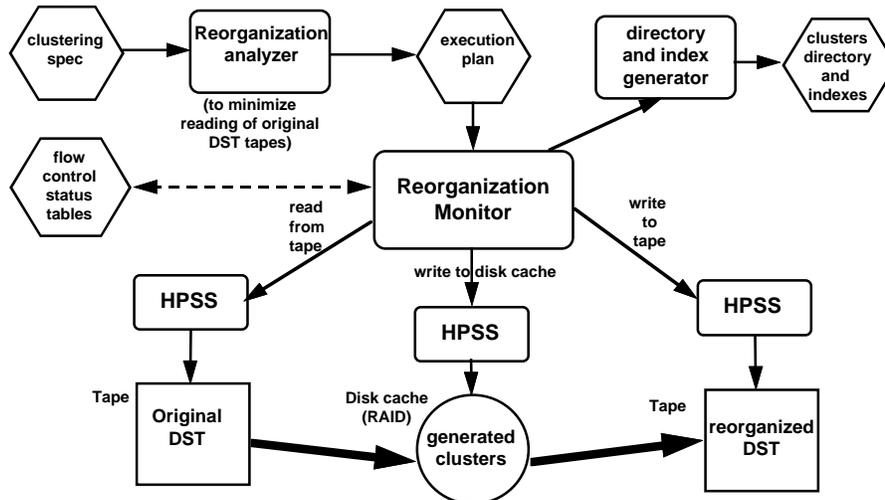
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- **Scenario A**
  - RAID more expensive than a Parallel Disk System (factor 2-3)
  - but, rely on HPSS to manage disk
  - storage management simplified
- **Scenario B**
  - a Parallel Disk System is cheaper, does not depend on RAID vendor
  - but, need to manage disk allocation
  - has control over placement of events on cache
- **Planned initial pilot**
  - Scenario A under NERSC

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## Reorganize DST tapes according to clusters

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## Open Problems

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- Discover event clusters in sparse high M-dim space given analyst guidance on binning
- Analyze the natural clustering of events by properties
- develop an efficient index on high M-dim space
- develop a cache management policy for job mixes
- Simulate and test the effect of distributing events on disk cache by blocks vs. one event per disk
- The benefit of partial event data replication to accommodate conflicting access patterns

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## Open Problems (cont'd)

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- **The ability of HPSS to store files on tapes according to external specifications**
- **The ability of HPSS to perform “partial file reads” from tape storage**
- **The possibility and effectiveness of parallel tape management under HPSS**
- **The benefit of partial event data replication to accommodate conflicting access patterns**